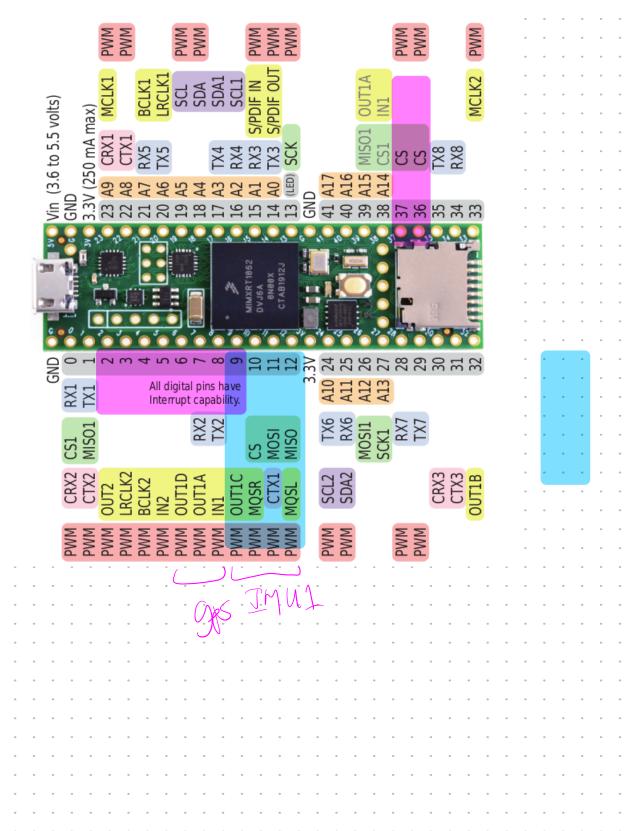


in conclusion: using the 2n7000 isn't going to work bc it doesn't pull the ckt high enough on the 5v output

## WRONG CONCLUSION

flip the source and drain --> the LV and HV sides and it works just fine! the problem from before was that the gate was NOT pulled high enough (translation: I didn't connect the gate correctly the first time)



## Extended Complementary Filter (with notes)

$$\frac{1}{6}q = \sum_{i=1}^{6} q_{i} \Delta t \qquad (1)$$

rate of change of q can be approximated using

$$\frac{1}{6}\hat{q} = \frac{1}{2} \frac{1}{6}\hat{q} \cdot \left(0 \left(w - ke\right)\right)^{T} (2) \qquad \text{we}$$

· error · calcs : -> 3-24

reference for gravity:

$$\overrightarrow{\mathcal{N}}_{r(\omega)} = (0 \ 0 \ -1) \ (3)$$

using estimates 
$$q = (q_w q_x q_y q_z)$$

$$\frac{1}{\sqrt{r}(\omega)} = \sqrt{\frac{1}{r}(\omega)} \sqrt{\frac{1}{r}(\omega)} = \frac{1}{r} \sqrt{\frac{1}{r}(\omega)}$$
(5)

$$R_{q}(\overline{q}) = \begin{pmatrix} q_{w}^{2} + q_{x}^{2} - q_{y}^{2} - q_{z}^{2} & \cdots & 2q_{x}q_{z} - 2q_{w}q_{y} \\ 2q_{y}q_{z} + 2q_{w}q_{x} \\ q_{w}^{2} - q_{x}^{2} - q_{y}^{2} + q_{z}^{2} \end{pmatrix} (7)$$

$$L_{V\Gamma(\vec{a})} = \begin{pmatrix} 2(q_{x}q_{z} - q_{w}q_{y}) \\ 2(q_{w}q_{x} + q_{y}q_{z}) \end{pmatrix}$$

$$(8)$$

$$2(q_{w}^{2} + q_{z}^{2}) - 1$$

$$q equestron$$

error  $\ell$  is calculated from two error terms. all vectors are normalized before calculations:

e measure 
$$\sqrt{\frac{1}{|V|}}$$
 (11)

reference  $\sqrt{\frac{1}{|V|}}$  (11)

 $C_a = \begin{pmatrix} x_0 \\ y_0 \\ 3\sigma \end{pmatrix} = V_{in}(a) \times V_{r}(a)$  (12)

$$\overrightarrow{V}_{\Gamma(m)} = ((V_{\Gamma \times} \cup \mathcal{O} \cup V_{\Gamma \times}))$$
 (13)

$$\nabla_{\Gamma}(e) = \left(0 \cdot 0 - 1\right) \times \left(V_{rx} \cdot 0 \cdot V_{rz}\right) \cdot (14)$$

$$\overrightarrow{\nabla}_{r(m)} = 2 \overrightarrow{\nabla}_{r(m)} q^{-1} \qquad (16)$$

$$\frac{1}{V_{r}(m)} = \frac{2(q \times q_{r}y + q_{u}q_{z})}{2(q_{x}^{2}y + q_{y}^{2}) - 1}$$
(17)

but it needs to change based on the conditions

$$\frac{1}{e} = \begin{cases}
\frac{1}{e} + \frac$$

apparently this can be conbined with the gyroscopic integration to provide with an accurate correction factor of gyroscopic drift.

## B. Universal convergence:

this mathemtical relationship can be described using the scalar version of this formula where

$$0 < \theta < \pi$$
 (radians)

(estimate from the next-timestep?)

$$|\vec{e}_a| = |\vec{v}_m(a)| \cdot |\vec{v}_r(a)| \sin\theta$$
 (22)

1. b/a they're normalized

$$|e_a| = \sin \theta$$
 (22a)  $\Rightarrow |e_a| = \theta$  for small  $|\theta|$  (22b)

since the gravity error term is normalized, the formula simplifies to the above formulas

also, the proper gain K must be smaller than 1 to get a stable convergence!

C. Magnetic Disturbance rejection

magnetic sources, whether ferromagnetic or diamegnetic, are everywhere, it distorts the overall signal where it is around 0.2 to 0.65 Gauss when considering all possible locations (though it can be very consistent on a local scale!)

what Madgwick et al does is reject magnitudes above or below this range!

$$\overline{V}_{m}(m) = \int \overline{V}_{m}(m) \cdot \int dr \cdot m_{mn} < |\overline{V}_{m}(m)| < m_{max}$$
 (23)

total error conditions, then, considering the above conditions:

$$\vec{e} = \begin{cases}
\vec{e}_{\alpha} + \vec{e}_{m} & |\vec{\nabla}_{m}(\vec{a})| > 0 & \text{kl. minn} < |\vec{\nabla}_{m}(\vec{n})| < m \text{ max} \\
\vec{e}_{\alpha} & \text{minn} < |\vec{\nabla}_{m}(\vec{n})| < m \text{ max}
\end{cases}$$
(24)

D. Magnetic Disturbance Compensation

there are a few reasons for why this is not an effective long-term solution. turning this off leaves the the IMU estimator susceptible to gyroscope drift. we have to add compensation. Keyframes are instances during a movement that define it and leaves interpolation to define the in-between frames.

Keyframes can be used to determine magnetic samples of minimal magnetic interference or gyroscopic drift.

this is achieved by calibrating hte gyroscopes and them performing the range of motions once.

quaternion q represents the movement due to accel and mag reference. so the accelerometer and magnetometer vectors that converge to q can be calculated by substituting 3 and 15 into 5 to produce the following values:

$$\overrightarrow{V}_{SIM(\vec{a})} \cdot k \overrightarrow{V}_{SZM(\vec{m})} \cdot \overrightarrow{T}_{SB} = (0 \ 0 \ -1)^{T} (3)$$

$$\overrightarrow{V}_{r(\vec{a})} = (0 \ -1 \ 0) (15)$$

$$\overrightarrow{V}_{TM(\vec{a})} \cdot (\vec{q}) = \vec{q} \cdot \vec{V}_{r(\vec{a})} \vec{q}^{-1} \cdot Q$$

$$\overrightarrow{V}_{SLM(\vec{m})} \cdot (\vec{q}) = \vec{q} \cdot \vec{V}_{r(\vec{m})} \vec{q}^{-1}$$

$$(25)$$

each keyframe therefore consists of these two 3D-vectors.

whne magnetometer is unavailable, accelerometer is compared against V\_sim(a). the keyframe with the smallest differences between the two frames is taken as the system estimation.

$$\frac{\hat{e}_{sim(m)} = \hat{V}_{m(\tilde{a})} \times \hat{V}_{sIn(m)} \times \hat{V}_{r(\tilde{m})}}{\hat{e}_{a} + \hat{e}_{m}} \times \hat{V}_{sin(m)} \times \hat{V}_{r(\tilde{m})} \times$$

Compare:

$$\overline{e}_{SIM(m)} = \widehat{V}_{M(\vec{a})} \times \widehat{V}_{M(\vec{m})} \times \overline{V}_{r(\vec{m})} \quad (18)$$

## E.

Gain K I guess K is able to change, in normal use, K will be low so the gyroscopic integration can be incorporated. See (2). it's defined as K\_norm. that's the K that's used in normal conditions, for initial convergence, it should be K\_init and should transition to K\_norm, the following is (28).

$$k(t) = \begin{cases} k_{\text{norm}} + \frac{t_{\text{init}} - t}{t_{\text{init}}} (k_{\text{init}} - k_{\text{norm}}) \\ k_{\text{norm}} \end{cases}$$

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 $K_norm$ ,  $K_init$ , and  $t_init$  are found by examining the behavior of the alg and tuned for specific effects. However, the values  $K_init = 10$ , provide fast convergence because the error is expected to be high; and  $K_norm = 10$  after  $t_init$  at 3 seconds provide good performance

